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## Citation

Akcakaya, Mehmet, Tamer A Basha, Warren J Manning, and Reza Nezafat. 2014. "Efficient calculation of g-factors for CG-SENSE in high dimensions: noise amplification in random undersampling." *Journal of Cardiovascular Magnetic Resonance* 16 (Suppl 1): W28. doi:10.1186/1532-429X-16-S1-W28. <http://dx.doi.org/10.1186/1532-429X-16-S1-W28>.

## Published Version

doi:10.1186/1532-429X-16-S1-W28

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# Efficient calculation of g-factors for CG-SENSE in high dimensions: noise amplification in random undersampling

Mehmet Akcakaya<sup>1\*</sup>, Tamer A Basha<sup>1</sup>, Warren J Manning<sup>1,2</sup>, Reza Nezafat<sup>1</sup>

From 17th Annual SCMR Scientific Sessions  
New Orleans, LA, USA. 16-19 January 2014

## Background

SENSE [1,2] is one of the most used parallel imaging techniques. In [1], uniform undersampling was employed to efficiently reconstruct an unalised image, whereas in [2], a conjugate gradient-based method (CG-SENSE) was used for reconstruction with arbitrary trajectories. SENSE framework allows the calculation of g-factors, characterizing the noise amplification for a given k-space trajectory and coil configuration [1]. However, calculation of g-factors for arbitrary trajectories in high dimensions is time-consuming [3]. Furthermore, noise characteristics of random undersampling, used in compressed sensing, is not well-understood. In this work, we use a Monte-Carlo (MC) method for fast calculation of g-factors for CG-SENSE similar to [4,5] and apply it to random Cartesian undersampling trajectories. **Theory:** SENSE involves a pre-whitening step [1,2], thus without loss of generality, we assume white noise. SENSE reconstruction solves  $\min_{\mathbf{m}} \|\mathbf{y} - \mathbf{E}\mathbf{m}\|_2$ , where  $\mathbf{E}$  is the system matrix, and  $\mathbf{y}$  are the undersampled measurements. The g-factor for the  $k^{\text{th}}$  voxel is given by  $g_k = \sqrt{(\mathbf{E}^*\mathbf{E})^{-1}_{k,k} [\mathbf{E}^*\mathbf{E}]_{k,k}}$ . Inverting  $\mathbf{E}^*\mathbf{E}$  is not feasible in high dimensions. Instead we note the  $g_k$  corresponds to the  $k^{\text{th}}$  diagonal of the reconstruction noise covariance matrix (for normalized coil sensitivities), where  $\mathbf{n}_{\text{recon}} = (\mathbf{E}^*\mathbf{E})^{-1}\mathbf{E}^*\mathbf{n}_{\text{meas}}$ , and  $\mathbf{n}_{\text{meas}}$  is measurement noise with identity covariance matrix. We calculate the sample correlation matrix using a MC approach (since sample mean goes to 0), as  $1/(p-1)\sum_p \mathbf{n}_{\text{recon}}^p (\mathbf{n}_{\text{recon}}^p)^*$  for  $p$  instances of  $\mathbf{n}_{\text{recon}}$ . Note we only calculate and store the diagonal elements of this matrix, significantly increasing efficiency.

## Methods

The MC method was first verified in a numerical simulation, where the g-factor was explicitly calculated for a 2D coil configuration, to determine how many MC simulations suffice. Whole-heart imaging was performed with an isotropic resolution of 1.3 mm using a 32-channel coil array. Two 4-fold accelerated acquisitions were performed, one with uniform undersampling ( $2 \times 2$  in the  $k_y$ - $k_z$  plane) and one with random undersampling. Coil sensitivity maps were exported. Images were reconstructed using SENSE (for uniform) and CG-SENSE (for both). g-factors were also calculated with the proposed approach.

## Results

Figure 1 shows the results of numerical simulations, indicating the method converges in ~50 MC simulations. Figure 2 shows the reconstructions associated with the two undersampling patterns and reconstructions, and the corresponding g-factors respectively. The results exhibit the semi-convergence property for random undersampling but not for uniform. Furthermore, the g-factor for random undersampling is smaller at its convergent point than for uniform.

## Conclusions

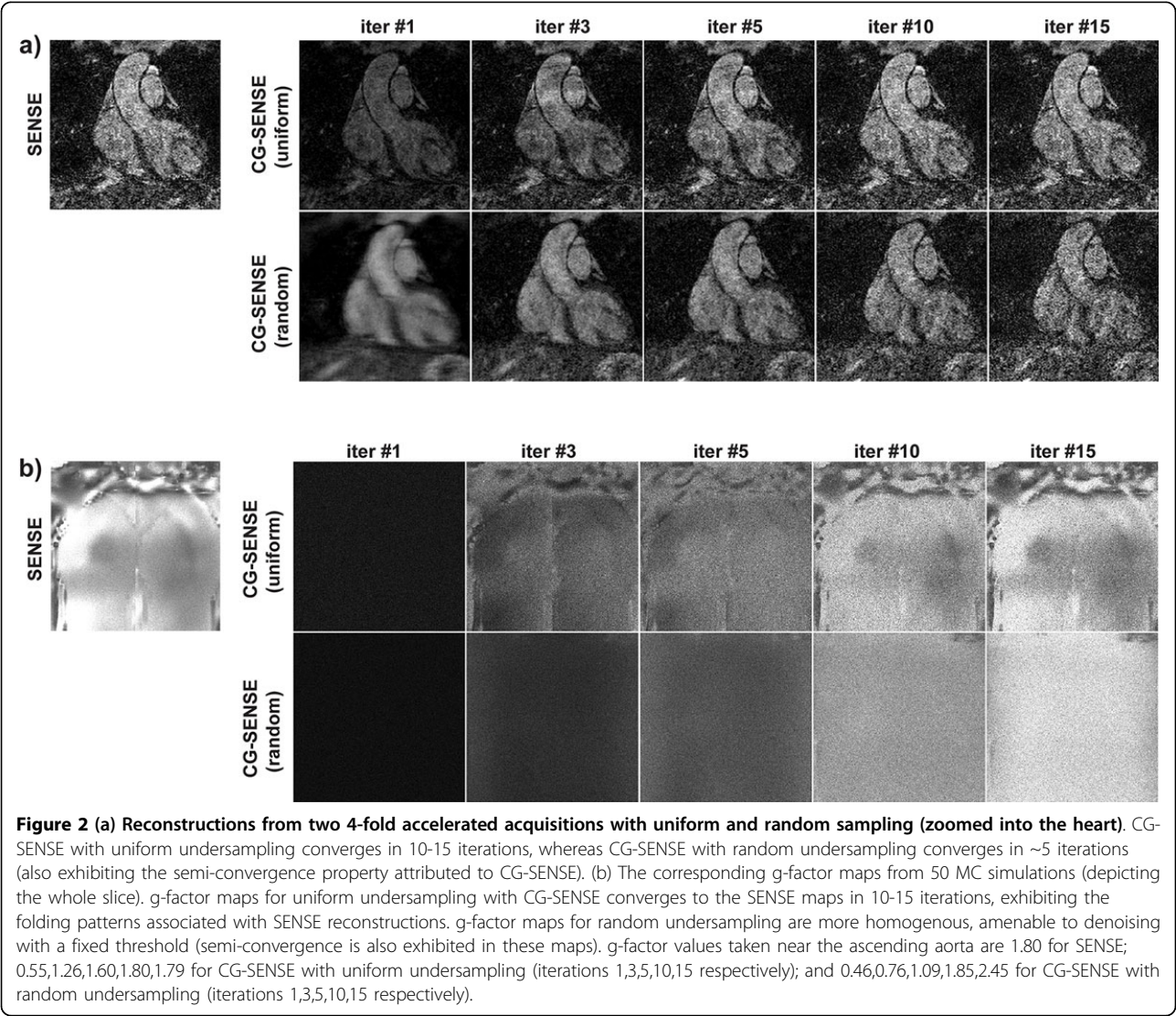
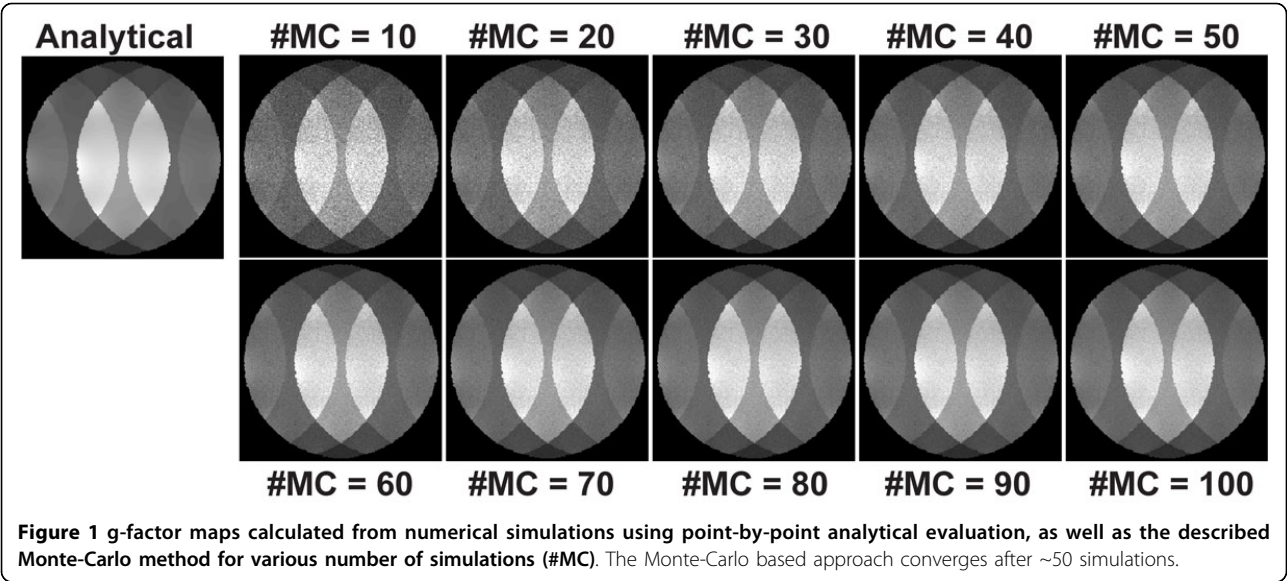
g-factors for random undersampling is better than those for uniform at high k-space dimensions and high acceleration rates.

## Funding

NIH:K99HL111410-01; R01EB008743-01A2.

<sup>1</sup>Medicine, Beth Israel Deaconess Medical Center, Harvard Medical School, Boston, Massachusetts, USA

Full list of author information is available at the end of the article



#### Authors' details

<sup>1</sup>Medicine, Beth Israel Deaconess Medical Center, Harvard Medical School, Boston, Massachusetts, USA. <sup>2</sup>Radiology, Beth Israel Deaconess Medical Center, Harvard Medical School, Boston, Massachusetts, USA.

Published: 16 January 2014

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doi:10.1186/1532-429X-16-S1-W28

**Cite this article as:** Akcakaya et al.: Efficient calculation of g-factors for CG-SENSE in high dimensions: noise amplification in random undersampling. *Journal of Cardiovascular Magnetic Resonance* 2014 16(Suppl 1):W28.

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